Computational Semantics and Knowledge Engineering

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User:

Mary has a dog.

Does Mary have an animal?

System:

User:

Mary has a dog.

Does Mary have an animal?

System:

Yes! (dogs are animals)

User:

Vincent is not married. Vincent is married.

System:

User:

Vincent is not married. Vincent is married.

System:

Hey! That's inconsistent.

User:

Florence is the cultural capital of Italy. Is Florence the capital of Italy?

System:

User:

Florence is the cultural capital of Italy. Is Florence the capital of Italy?

System:

Yes! (a cultural capital is a capital)

User:

Mia's husband is a gangster. Is Mia married?

System:

User:

Mia's husband is a gangster. Is Mia married?

System:

Uhh, don't know...

Knowledge acquisition

- You need knowledge?
- You want a machine to get it?
- Why don't you get it from ... *text...*?
 - Wikipedia
 - Definitions
 - Dictionary glosses

Catch 22

- To build an intelligent NLP system we need background knowledge
- To acquire background knowledge automatically we need NLP



This talk

- Background: computational semantics
- Building a text understanding system
 - Syntax-semantics interface
 - Semantic representation and inference
 - Knowledge
- Open-domain question answering
 - Why you need semantics
 - Why you need knowledge
- A case study

Computational Semantics

Semantics

The branch of logic and linguistics concerned with meaning

Computational Semantics
 Using a computer to build meaning
 representations, and reason with the
 result (inference)

Applications

- Spoken Dialogue Systems
- Question-Answering Systems
- Textual Inference Systems





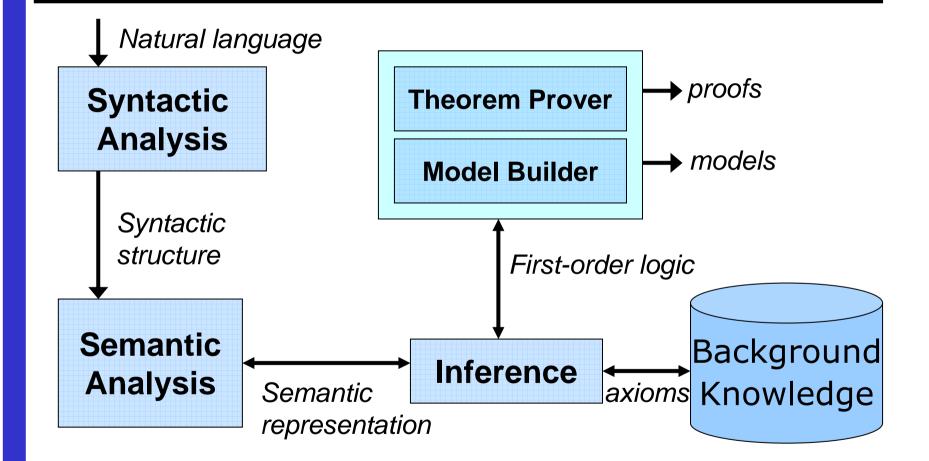
Reasoning and Natural Language

- To reason you need
 - A logical formalism
 - A reasoning engine for your logic
 - Something that maps language into logic
 - Supporting background knowledge

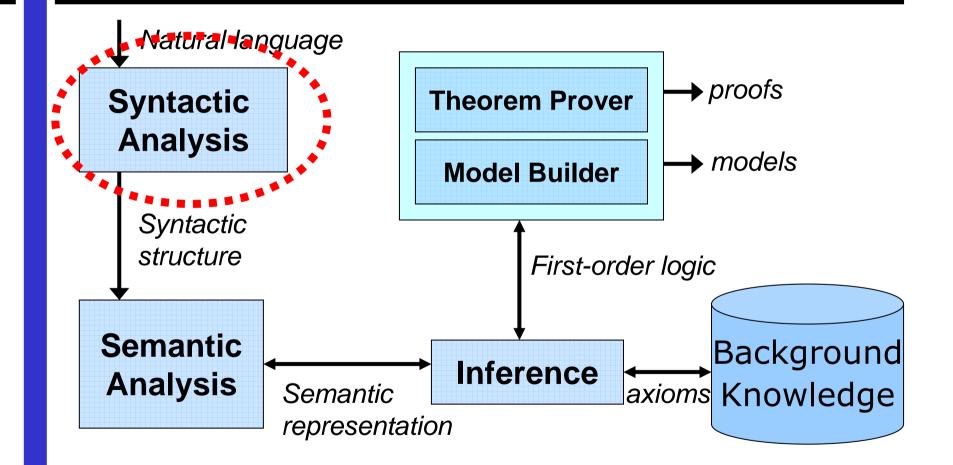
Reasoning and Natural Language

- To map language into logic you need
 - A parser that produces syntactic structure
 - A syntax-semantics interface
 - A semantic formalism

The bigger picture



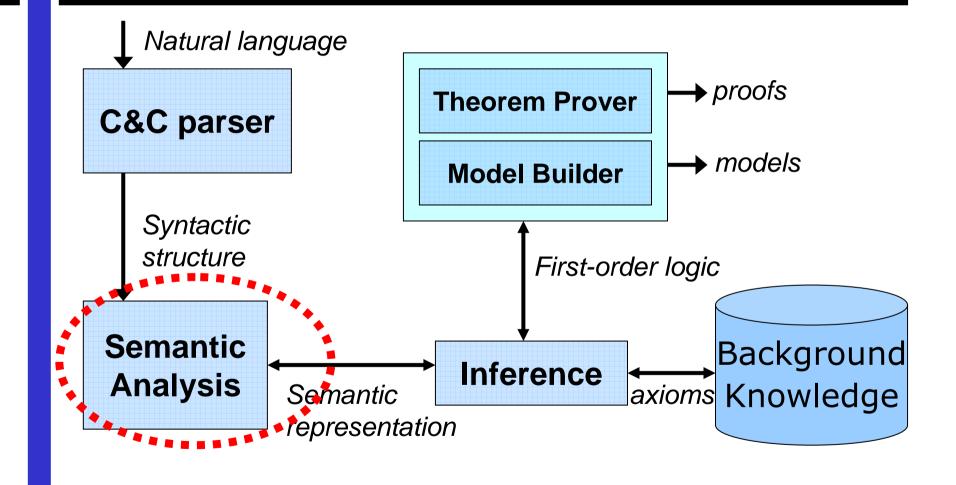
The bigger picture



Syntactic analysis: C&C

- Creation of tree-banks
- Stochastic parsers trained on such tree-banks
- C&C parser (Clark & Curran)
 - Combinatory Categorial Grammar
 - Efficient and robust

The bigger picture



Semantic Analysis: Boxer

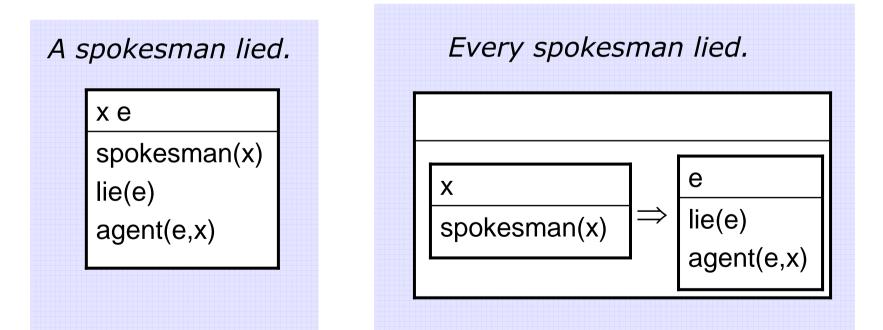
- The recent developments in parsing has lead to developing wide-coverage semantic analysis components
- One of such systems is Boxer, developed over the last five years



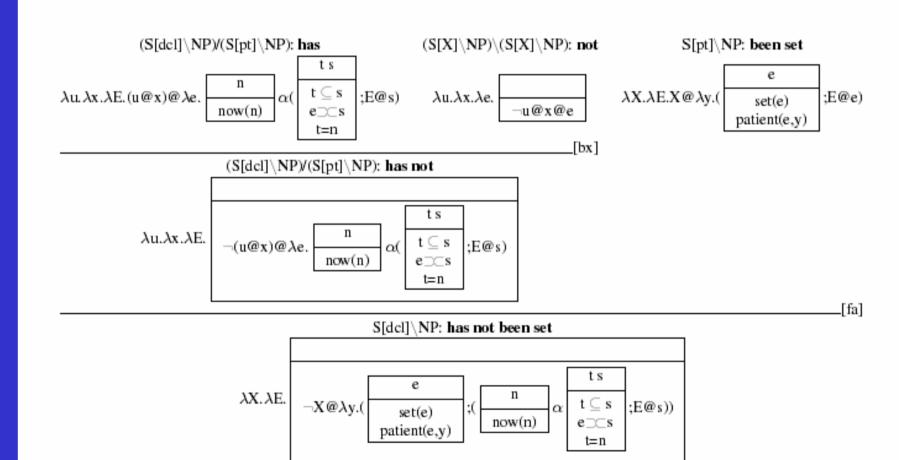
- Follows the principles of Hans Kamp's
 Discourse Representation Theory
- Produces formal semantic representations
- Translation to first-order logic
- Systematic syntax-semantic interface, using lambda calculus
- Pronoun and presupposition resolution

Discourse Representation Theory

- Box-like structures as semantic representation
- Structure plays role in pronoun resolution
- Neo-Davidsonian event semantics [VerbNet]

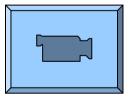


Syntax-semantics interface

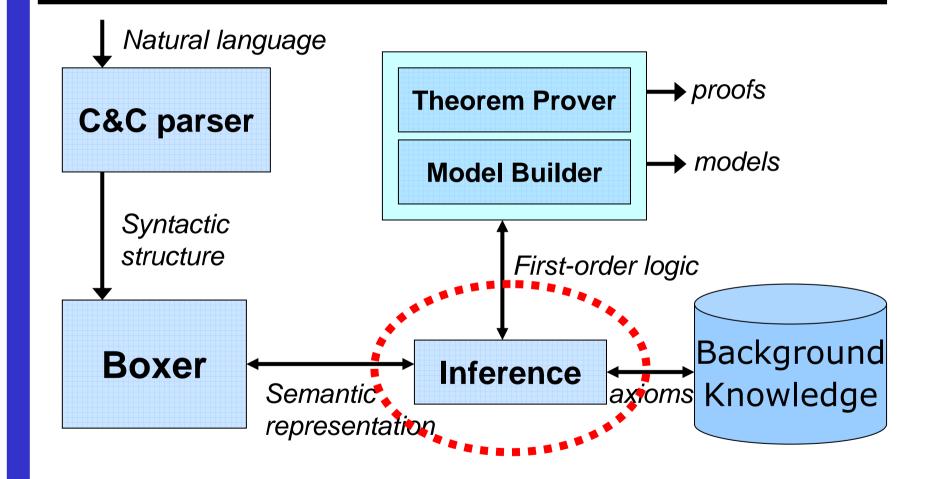


Breakthrough

- Why C&C and Boxer make a difference
 - Broad-coverage
 - Reasonably efficient
 - Clean syntax-semantics interface
 - Interpretable structures



The bigger picture

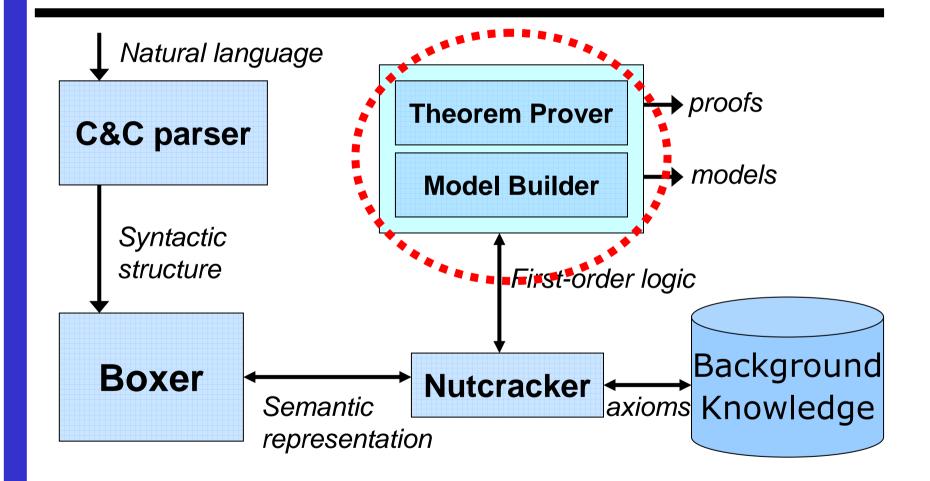


Inference: Nutcracker



- Selects appropriate background knowledge for inference
- Acts as a mediator between Boxer and inference engines for first-order logic
 - Theorem prover
 - (Finite) Model builders
- Reports results back to Boxer or frontend application

The bigger picture



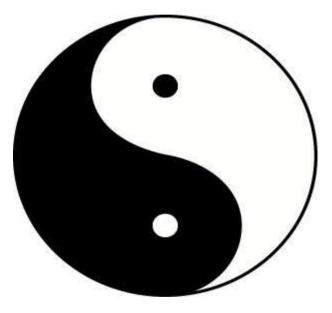
Inference engines

- Which inference engines? Off-the-shelf!
- How do we know which are the best?
 - CADE world cup automated deduction
 - Theorem proving: vampire
 - Model building: paradox

Some concerns

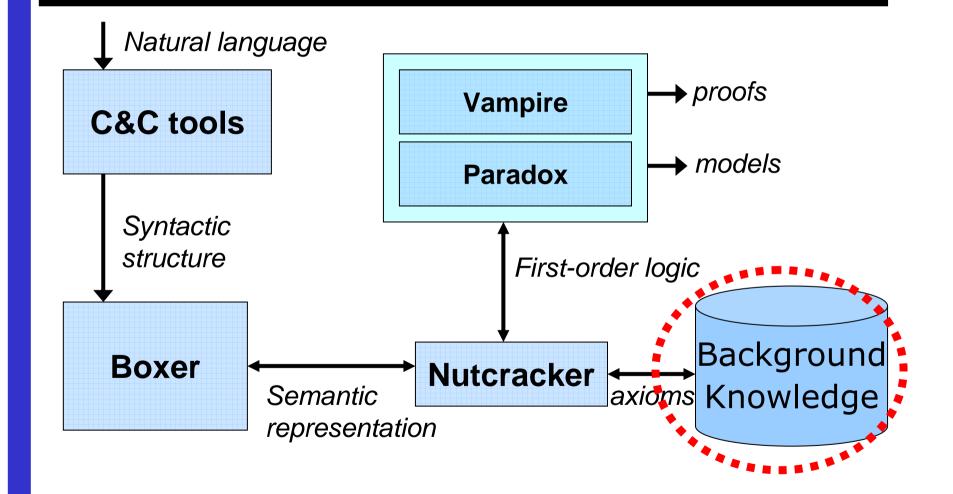
- Isn't first-order logic undecidable?
- Why do we need both a theorem prover and a model builder?

Yin and Yang of Inference



Theorem Proving and Model Building function as opposite forces

The bigger picture



Background Knowledge

 Format: first-order logic

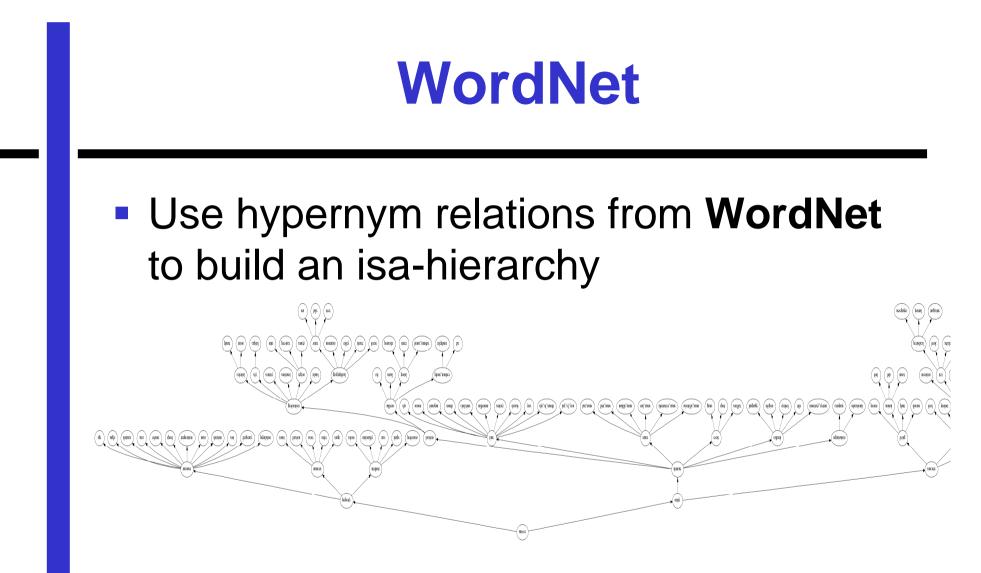
 Amount: ideally as few axioms as possible

Man or Machine?

Manually constructed

- WordNet
- NomLex
- CIA Factbook
- CYC/OpenCYC
- Automatically constructed
 - Linguistic patterns [e.g. Hearst]
 - Paraphrases [e.g. Lin & Pantel]





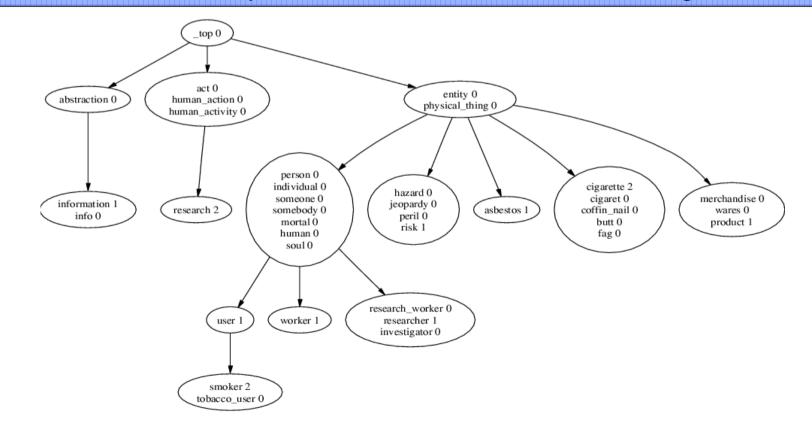
Create MiniWordNets for small texts

Convert these into first-order axioms

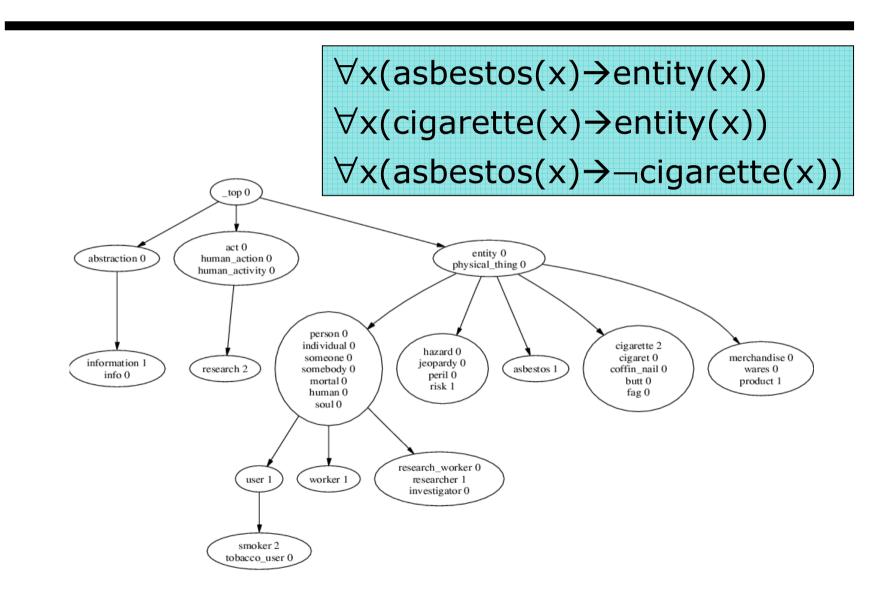
MiniWordNet

There is no asbestos in our products now.

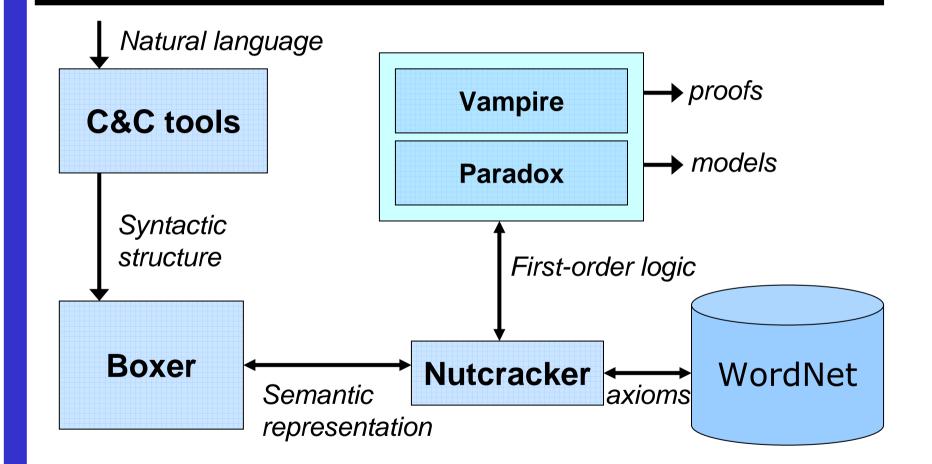
Neither Lorillard nor the researchers who studied the workers were aware of any research on smokers of the Kent cigarettes.



MiniWordNet



The bigger picture



OK – does this work?

It does work for small domains...

Examples:

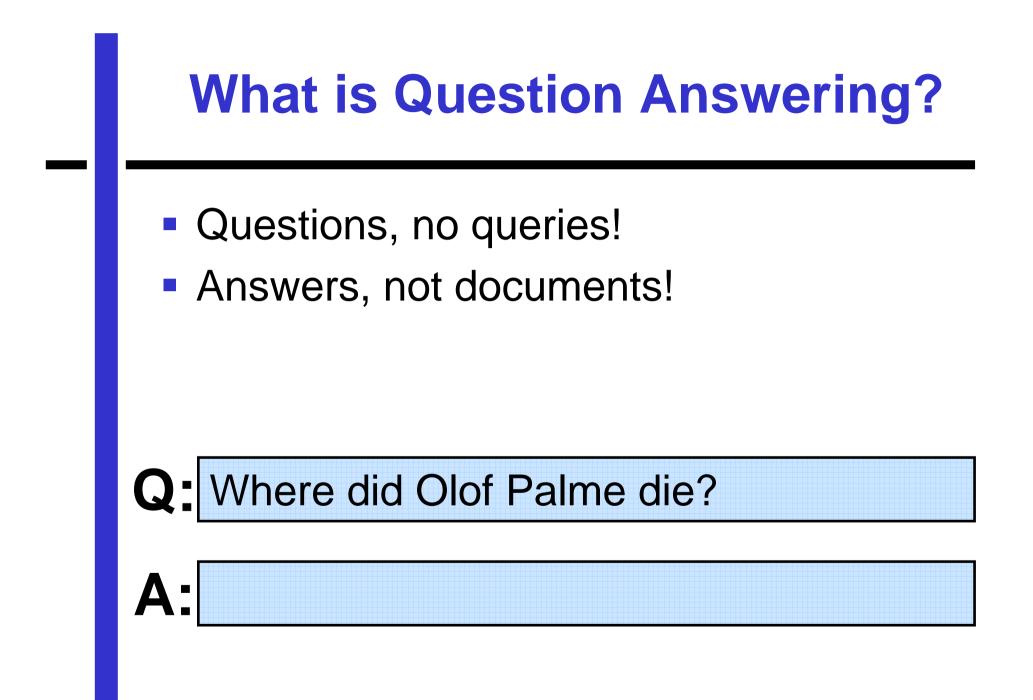
- Talking robots
- Smart houses

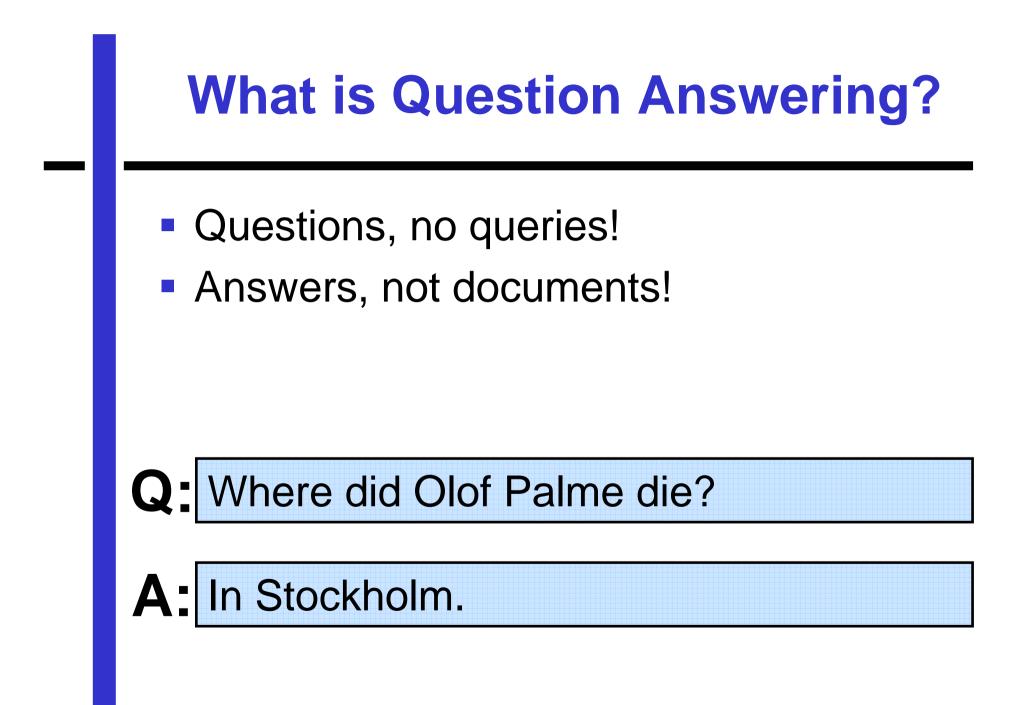




But what about larger domains?

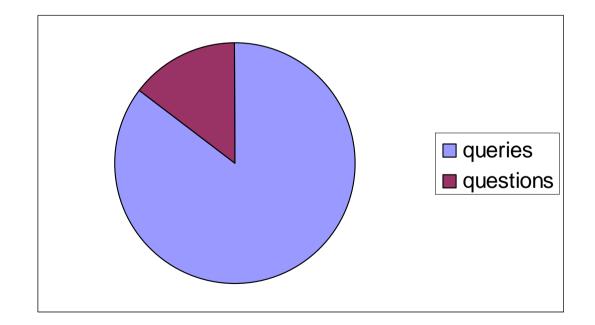
- Semantic interpretation of e.g.
 - newspaper texts
 - web pages
 - wikipedia
- Let's look at a real-world application in which we used C&C and Boxer:
 - Open-domain question answering
 - NIST/TREC campaigns
 - Pronto QA system





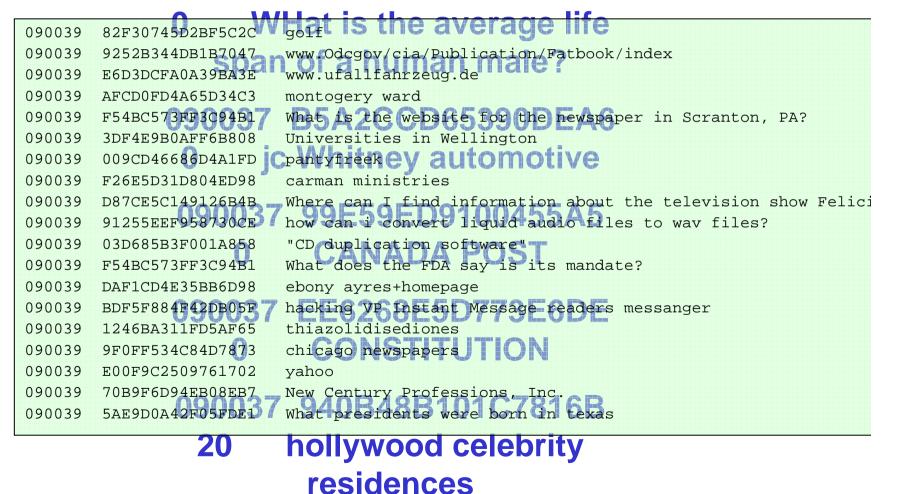
People ask questions

- Excite Search Engine Log with 2,477,283 queries
- Collected at 20 December 1999
- Around 15% natural language questions



10 arrowsmith 095057CIL9596459409 10 Wrestlemania 2000

090037 E0879EB63392B1FA

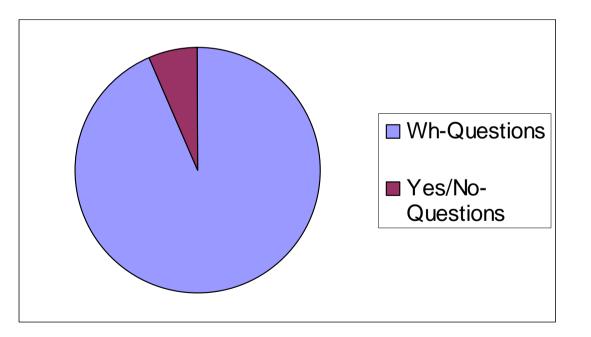


Question Types

Wh-Questions:

Yes/No-Questions:

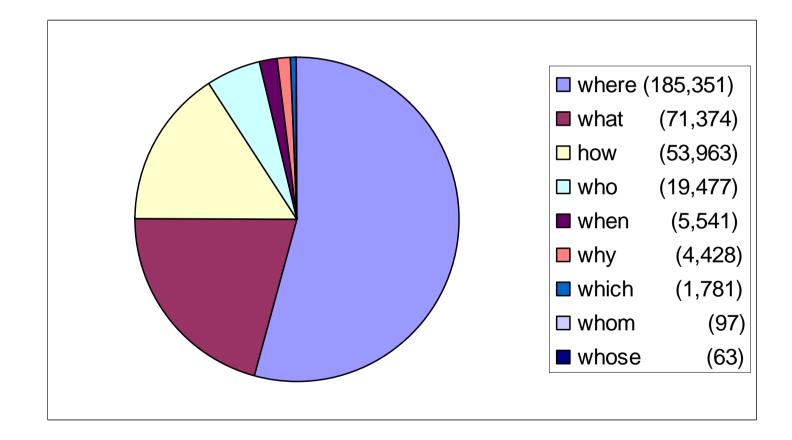
342,075 23,882



Question Types

How is the U.N. funded? Was Don lying about the shooting? Who invened the rangefinder? what exactly is a hydraulic system? When was Fisher Price started? Where can i find exams on the internet about database management systems course with sol Where can I find yesturday's weather? What is the probability of extraterrestrial life existing in the Universe? What vacation rentals are available in the pacific Northwest? Who was president Grant's private secretary? Where can I get a recipe for sloppy joes? What are the times of suprises and supsets in various cities in the US? pros and cons of professional managers in government? Where can I find information about FLSA? where can I find details of property in Scotland? Who sings the song "Ice, Ice Baby"? what percentage of schools have internet access? When were gingerbread houses invented? how do I find out the purchase price of a house that sold in California? Are cmputer chips made from sand? what is lupus? where can i find how to write a resume? WHERE CAN I FIND ON-LINE GREETING CARDS?

Wh-Question Types



Natural search

- Google gets about 200 million searches per day
- That would mean about 30 million natural language searches

Trying to guess

Q: What country is Berlin in?

A: Ties between **Germany** and Iran strained after a <u>Berlin</u> court verdict ruled on April 10 that Iran's top leaders were behind the assassination of four exiled Kurdish opposition leaders in a Berlin restaurant.

Trying to guess

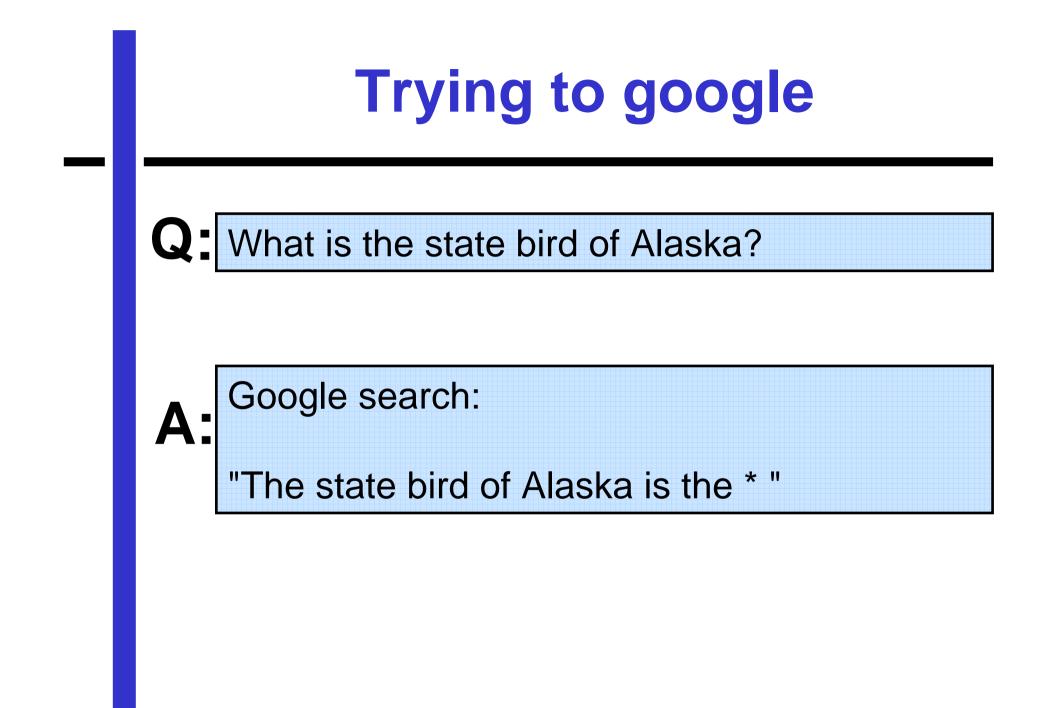
Q: What country is Berlin in?

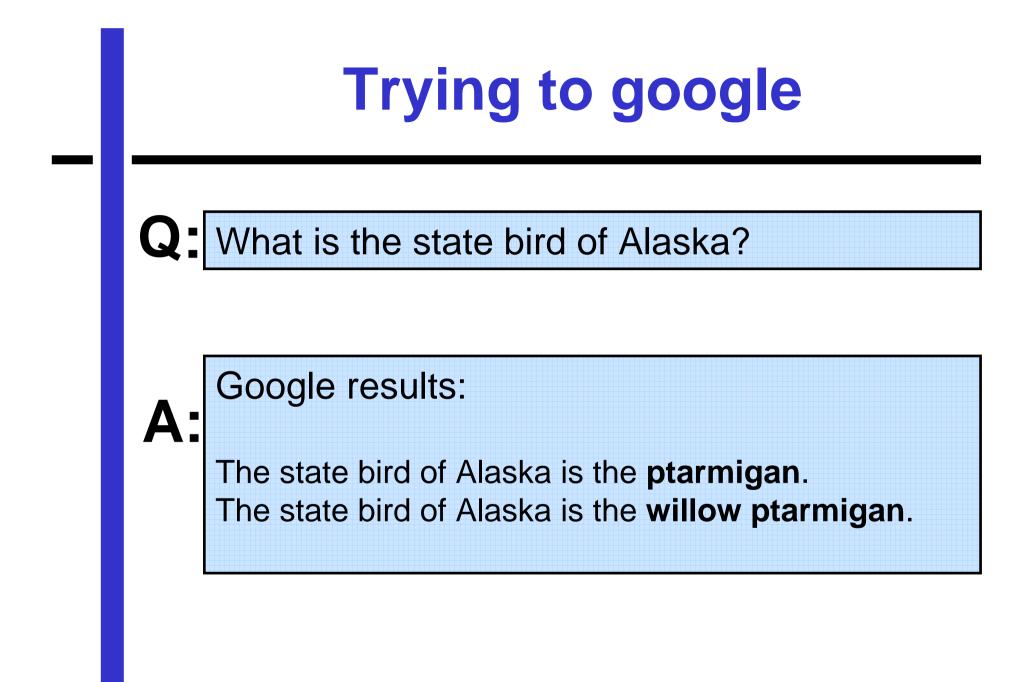
A: Ties between Germany and Iran strained after a <u>Berlin</u> court verdict ruled on April 10 that Iran's top leaders were behind the assassination of four exiled Kurdish opposition leaders in a Berlin restaurant.

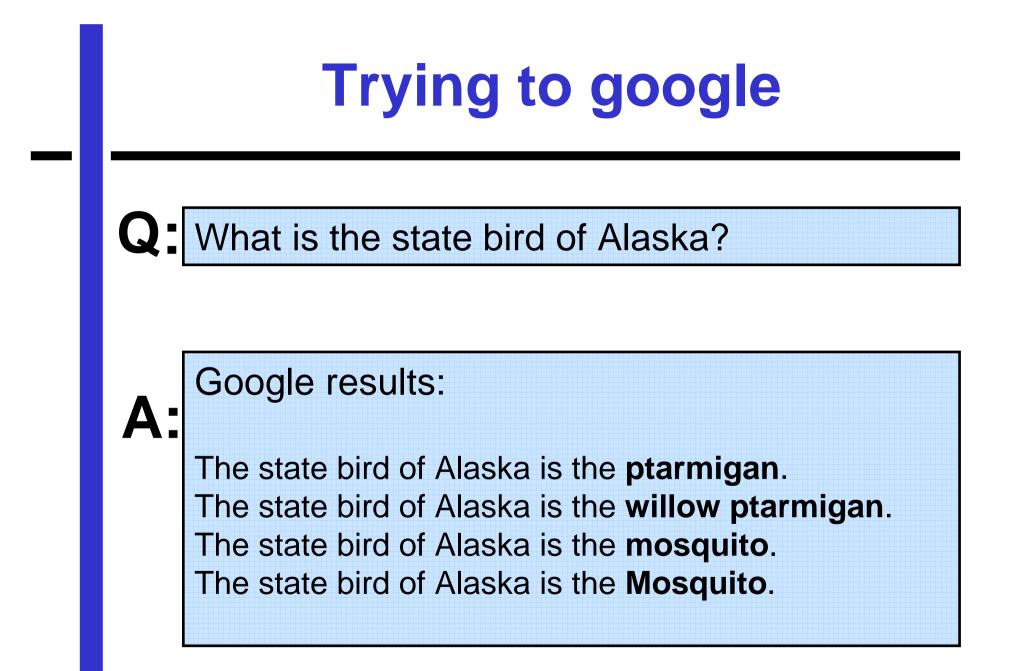
Another guessing example

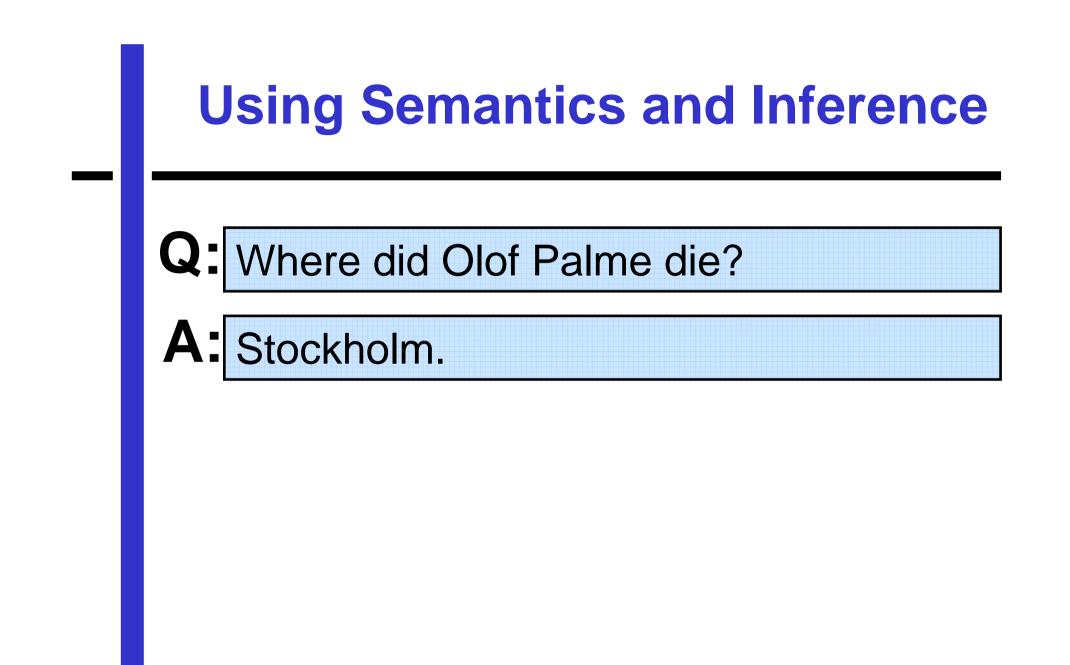
Port Arthur Massacre.What was the killer's nationality?

A nation asks why, the portrait of a lone gunman, Martin Bryant, should have no reason to be a <u>killer</u>. The man arrested after the <u>Port Arthur</u> massacre should be a wealthy man, only four years ago he inherited more than 500,000 **Australian** dollars 375,000 US dollars from one of the heirs of George Adams's great Tattersalls fortune.









Inference, internally

Where did Olof Palme die?

A:

given clause #13: (wt=2) 69 [hyper,47,16] vehicle(\$c3). given clause #14: (wt=2) 71 [hyper,50,16] vehicle(\$c4). given clause #15: (wt=2) 74 [hyper,52,18] building(\$c5). given clause #16: (wt=3) 51 [] have(vincent,\$c4). given clause #17: (wt=2) 76 [hyper,55,18] building(\$c6). given clause #18: (wt=2) 78 [hyper,61,16] vehicle(\$c6). given clause #19: (wt=2) 80 [hyper,65,10] organism(\$c1). given clause #20: (wt=2) 84 [hyper,67,10] organism(\$c2). given clause #21: (wt=3) 53 [] die(palme,\$c5). given clause #22: (wt=2) 86 [hyper,69,15] instrument(\$c3). given clause #23: (wt=2) 88 [hyper,71,15] instrument(\$c4). given clause #24: (wt=2) 90 [hyper,74,7] artifact(\$c5). given clause #25: (wt=2) 94 [hyper,76,7] artifact(\$c6). given clause #26: (wt=3) 56 [] \$c7=\$c6. given clause #27: (wt=2) 96 [hyper, 78, 15] stockholm(\$c6). -----> EMPTY CLAUSE at 0.01 sec ----> 113 [hyper,96,24,76] \$F.

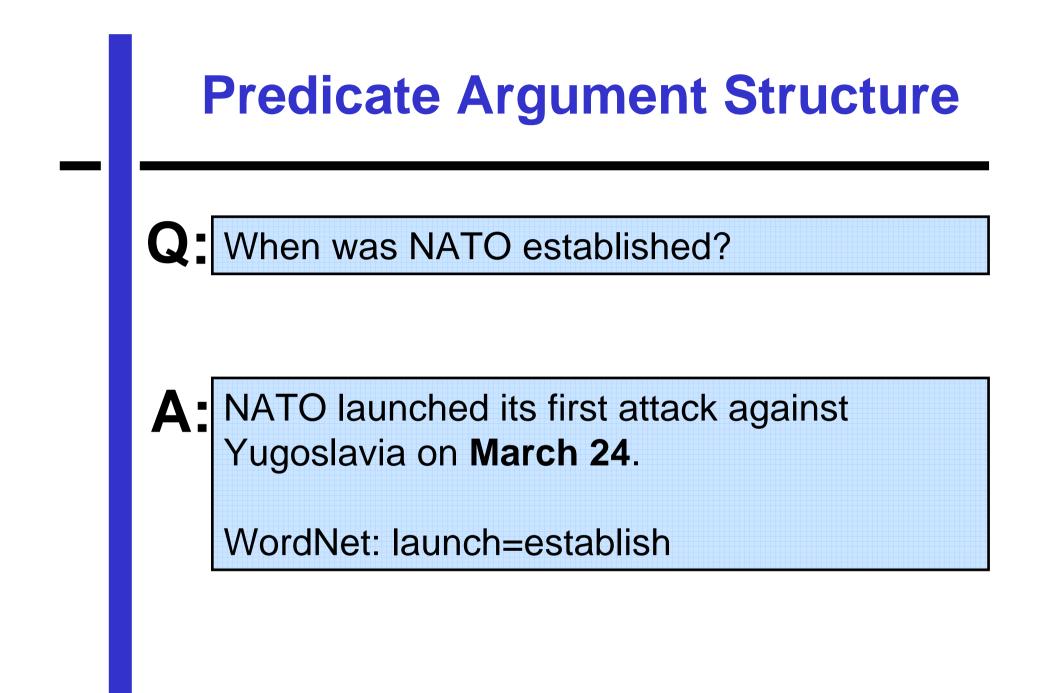
Inference, externally

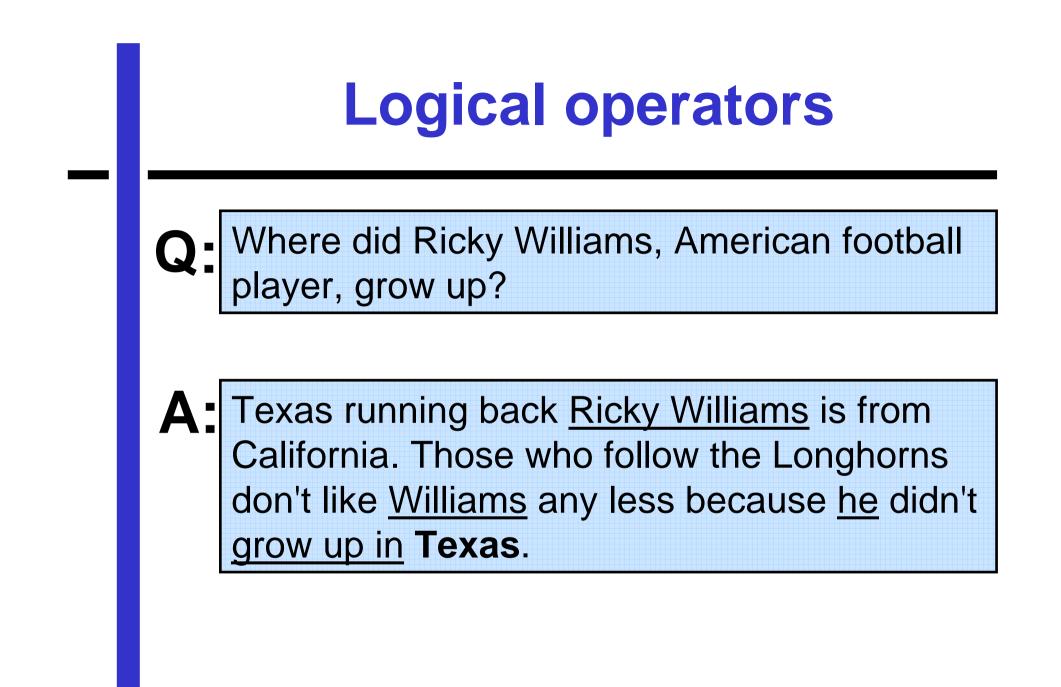
Q: Where did Olof Palme die?

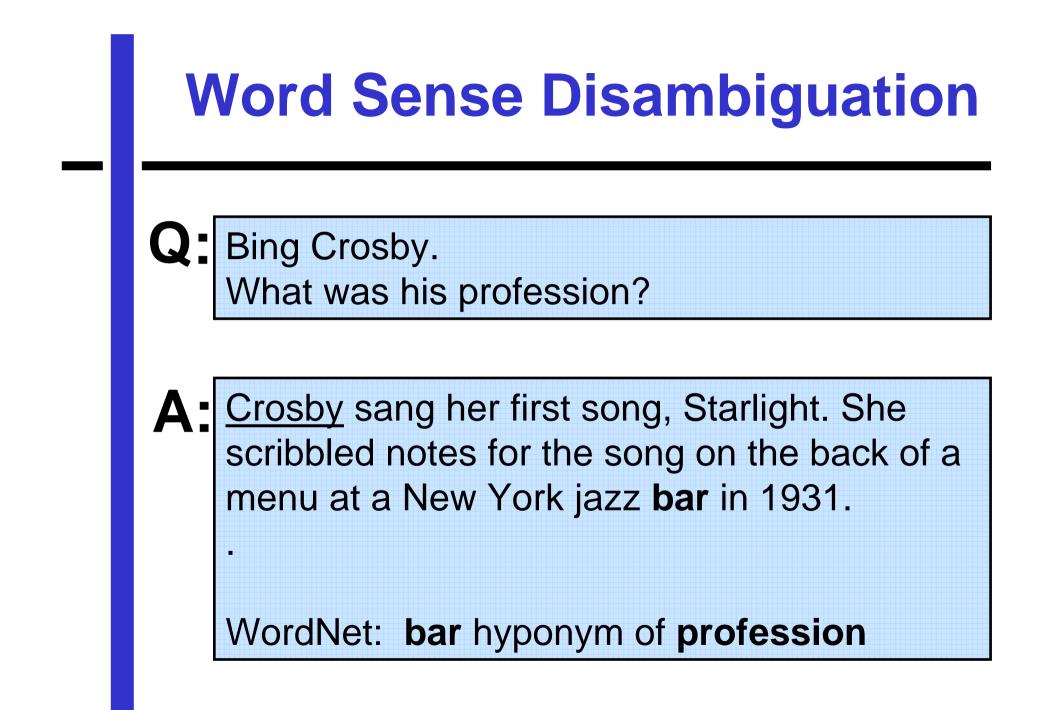
- 1. You are looking for location.
 - 2. I know that Stockholm is a city.
 - 3. Every city is a location.
 - 4. If x is shot to death then x died.
 - 5. I found the following evidence in document APW20000227.0124:
 - "In 1986, Swedish Prime Minister Olof Palme was shot to death in central Stockholm. "
 - 6. Hence Stockholm is the answer.

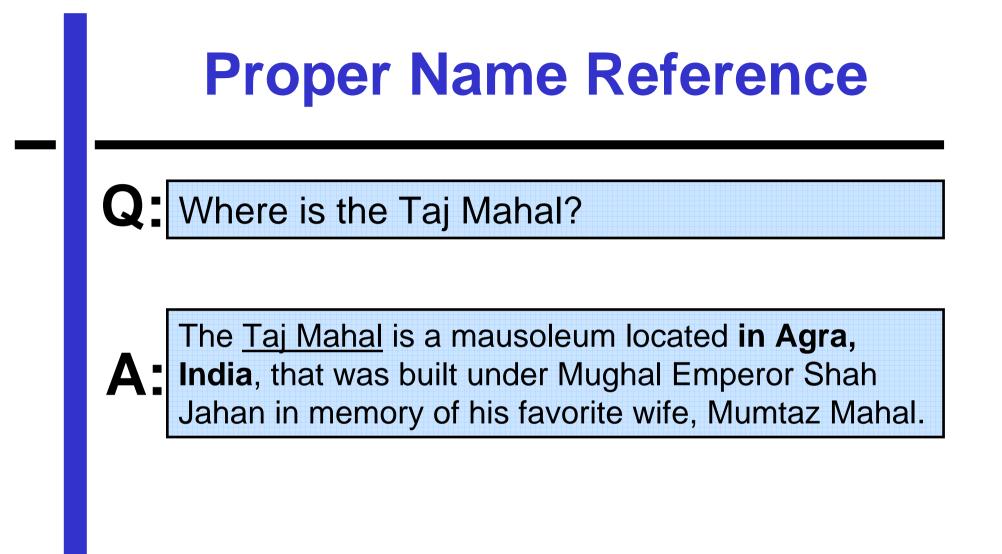
Why bother?

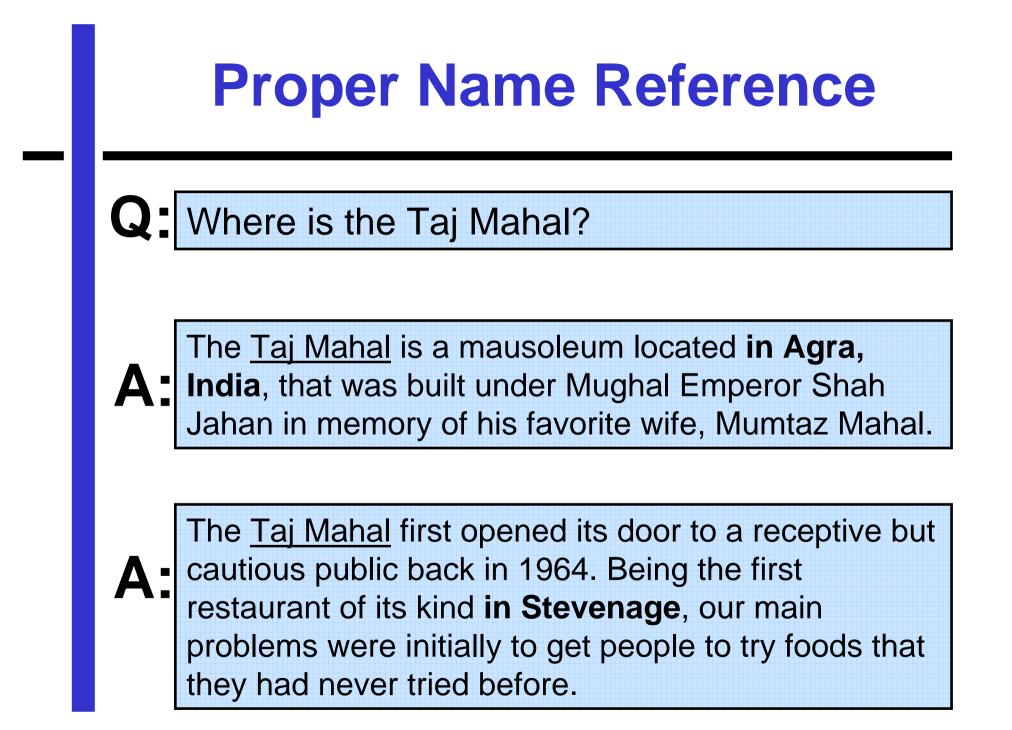
- Why Semantics?
- Consider some examples [not made up!]

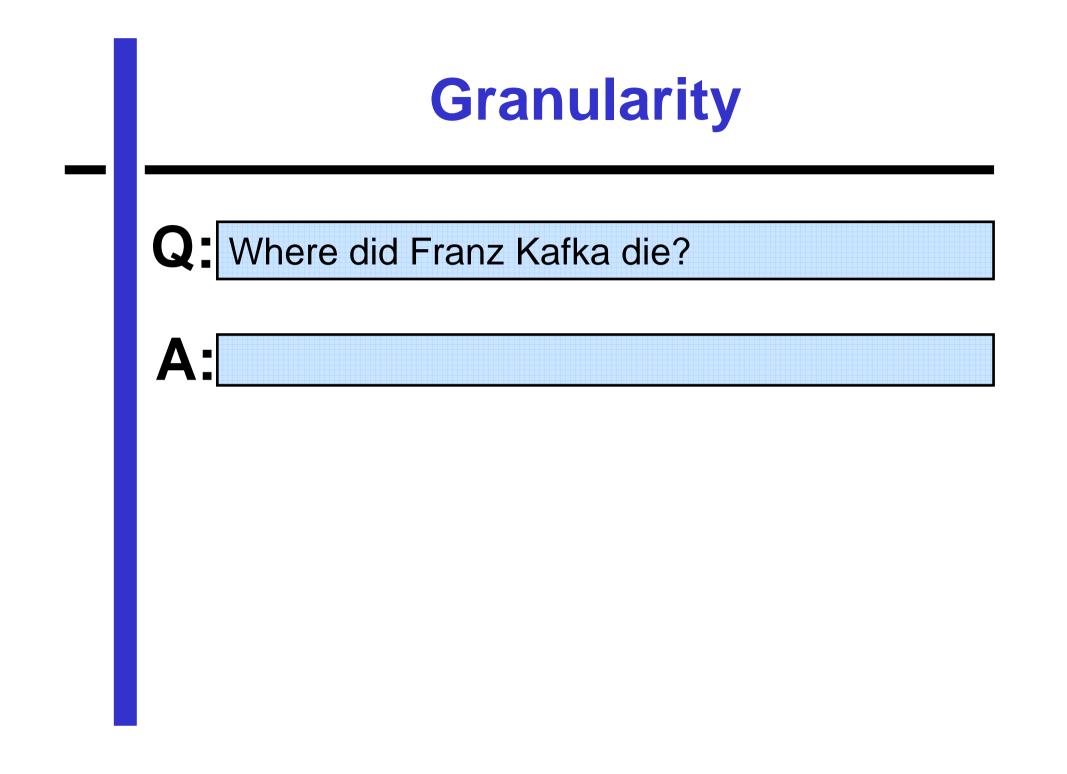


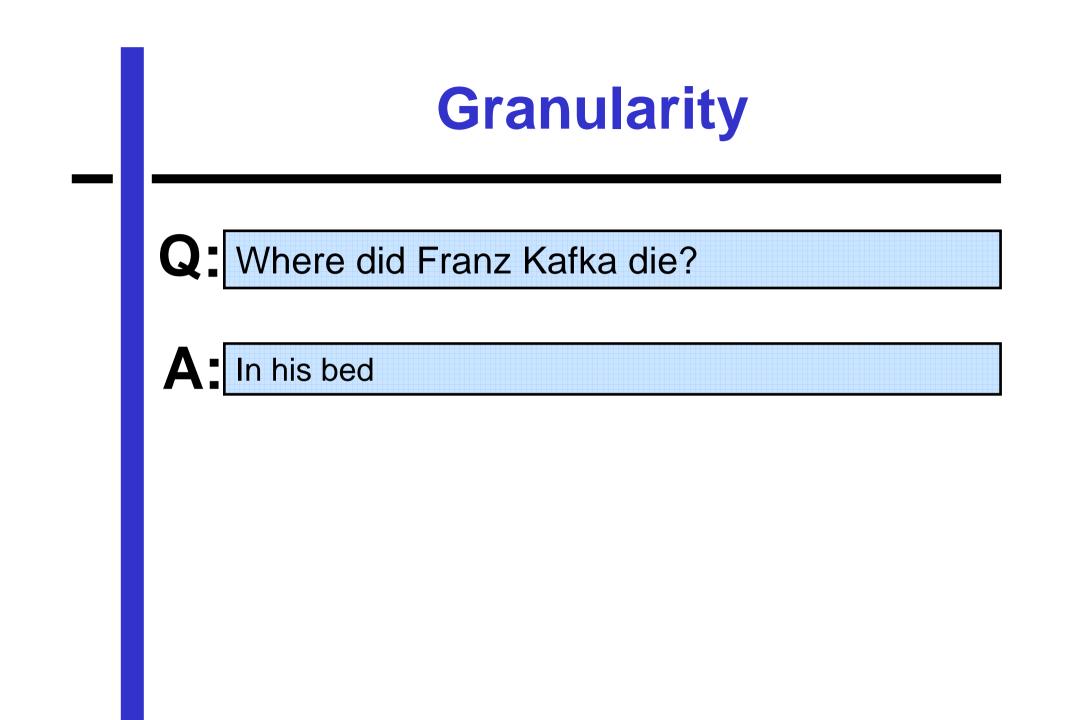


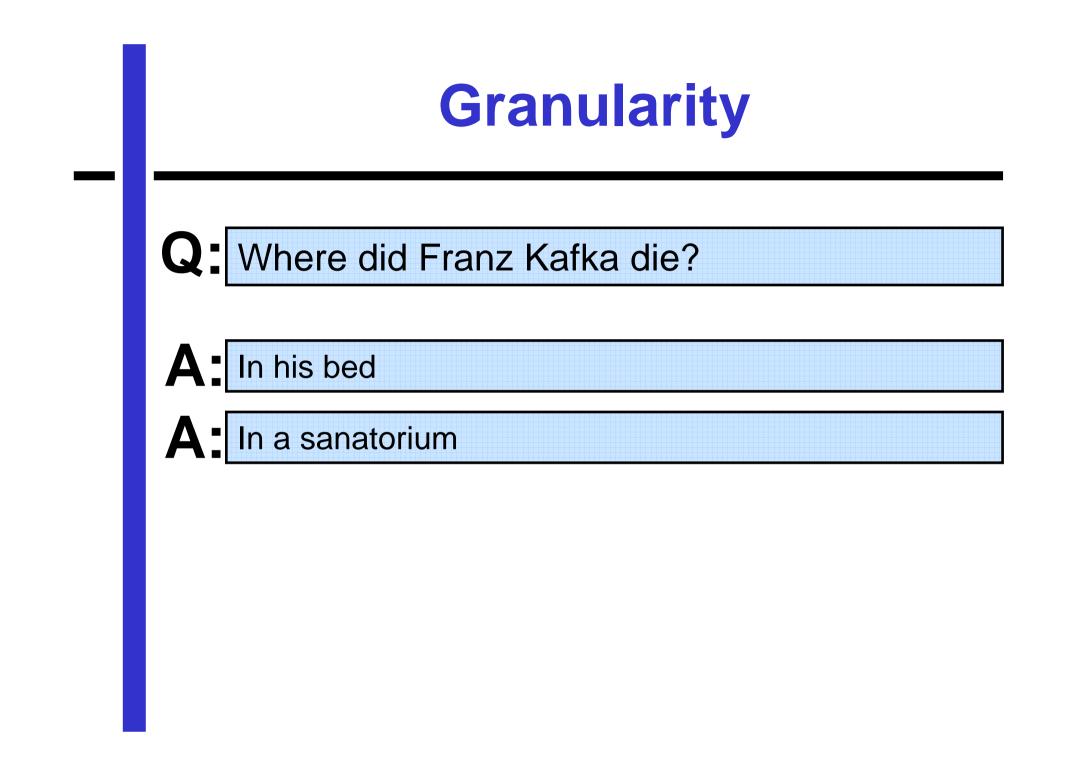


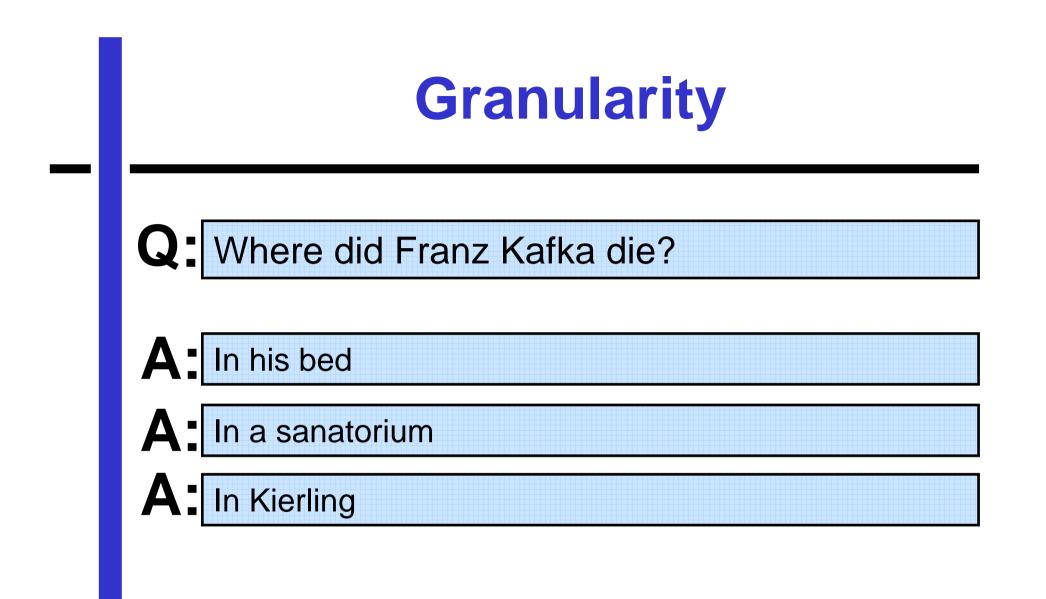


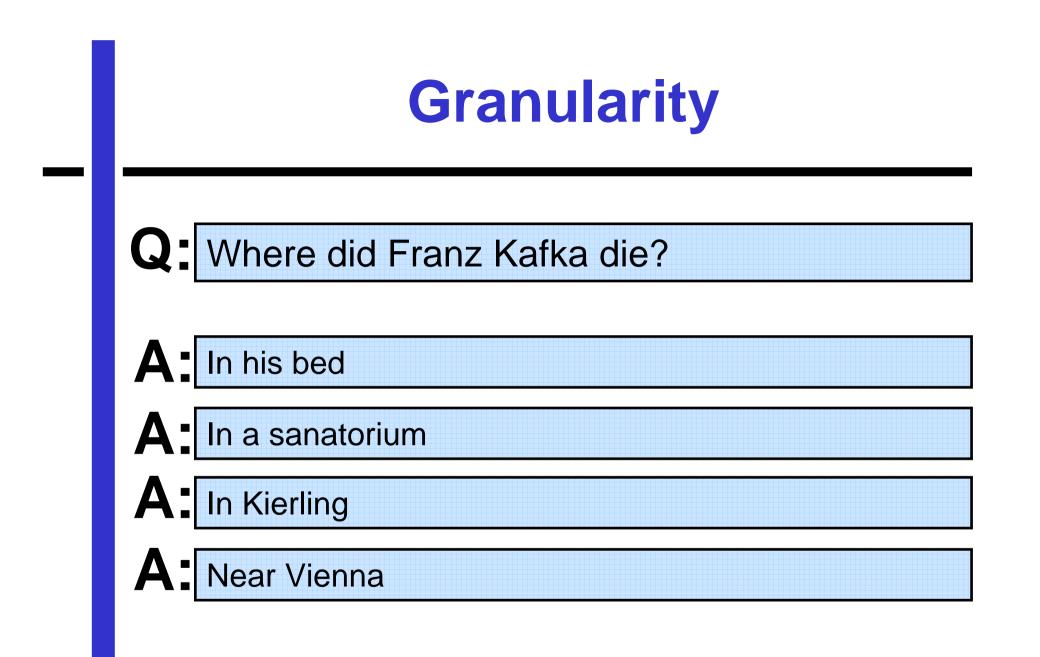


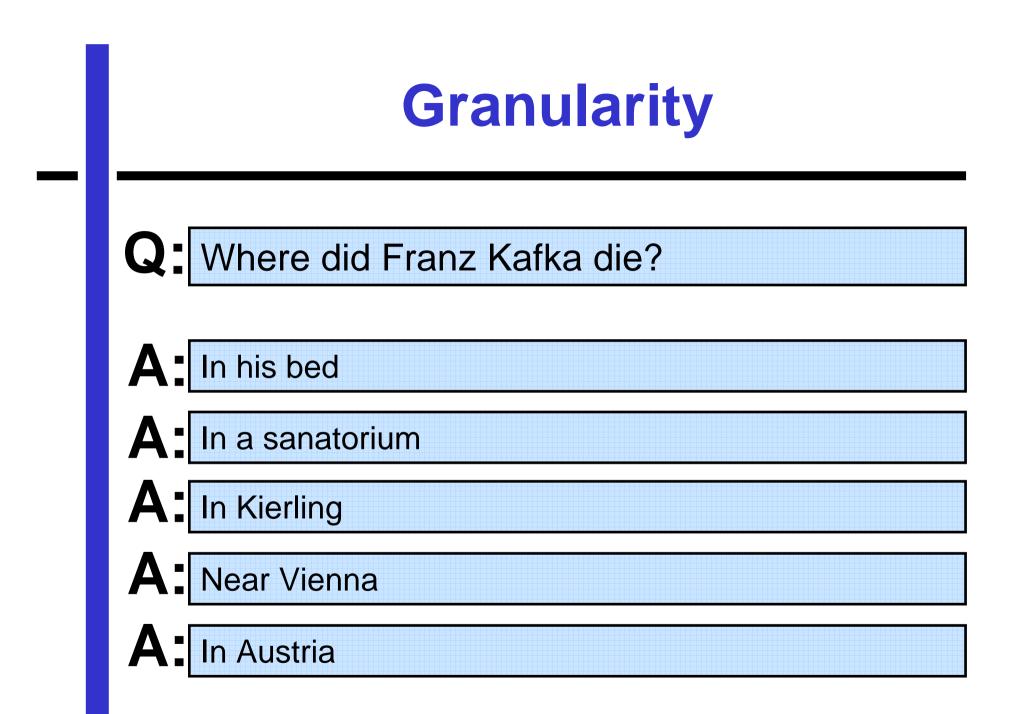












Moral

 Perhaps semantics and inference will get you quite far, but there is still this need for knowledge

Knowledge acquisition

- What kind of knowledge do we need for QA applications (apart from WordNet)
- Can we produce this on the fly?
- Let's have a look at some examples
 - Linguistic patterns
 - Paraphrase collections

Failing instances

TREC 20.2 (Concorde) What airlines have Concorde in their fleets?

WordNet has no instances of airlines.

• Note: DBpedia does. So does CYC.

Find instances with patterns

TREC 20.2 (Concorde) What airlines have Concorde in their fleets?

- Search for linguistic patterns in corpora (Hearst 1992)
- Example pattern: "X such as Y and"
- Text: ... said that airlines such as Continental and United now fly...

Find instances with patterns

TREC 20.2 (Concorde) What airlines have Concorde in their fleets?

- Search for linguistic patterns in corpora (Hearst 1992, Aguado de Cea et al. 2008)
- Example pattern: "X such as Y and"
- Text: ... said that airlines such as Continental and United now fly...

Pattern result

TREC 20.2 (Concorde)

What airlines have Concorde in their fleets?

Knowledge (Acquaint corpus):

Air Asia, Air Canada, Air France, Air Mandalay, Air Zimbabwe, Alaska, Aloha, American Airlines, Angel Airlines, Ansett, Asiana, Bangkok Airways, Belgian Carrier Sabena, British Airways, Canadian, Cathay Pacific, China Eastern Airlines, China Xinhua Airlines, Continental, Garuda, Japan Airlines, Korean Air, Lai, Lao Aviation, Lufthansa, Malaysia Airlines, Maylasian Airlines, Midway, Northwest, Orient Thai Airlines, Qantas, Seage Air, Shanghai Airlines, Singapore Airlines, Skymark Airlines Co., South Africa, Swiss Air, US Airways, United, Virgin, Yangon Airways

Paraphrases

TREC 4.2 (James Dean) When did James Dean die?

APW19990929.0165: In **1955**, actor James Dean was killed in a two-car collision near Cholame, Calif.

DIRT database (Lin & Pantel):

"X was killed in Y" == "X died in Y"

Paraphrases

TREC 4.2 (James Dean) When did James Dean die?

APW19990929.0165: In **1955**, actor James Dean was killed in a two-car collision near Cholame, Calif.

In first-order logic:

 $\forall x \forall t (\exists e(kill(e) \& theme(e, x) \& in(e, t))) \rightarrow \exists e'(die(e') \& agent(e', x) \& in(e', t)))$

General Knowledge

TREC 14.4 (Horus) Who was his father?

XIE19990713.004: It also hosted statues of Amon's wife, Mut, the goddess Isis, her husband, **Osiris**, and their son Horus.

Knowledge (manually coded?):

 $\forall x(husband(x) \rightarrow male(x)) \\ \forall x \forall y((son(x,y)\&male(y)) \rightarrow father(y,x)) \end{cases}$

Knowledge in QA

- We need knowledge for inferencebased QA system
- This is knowledge not explicitly expressed in the text or question
- Current background knowledge resources are not sufficient
- Automatically extracted knowledge
 - Ideally word sense disambiguated
 - Deal with more complex relations

Automatically harvested knowledge

- Good at simple relations
 - hyponyms, instances, synonyms
 - typically high precision and low recall
- Not so good at paraphrases
 - DIRT paraphrases [Lin and Pantel]
 - no improvement on QA and RTE
 - typically low precision, high recall
- General knowledge is hard to get, but of course we can use Boxer itself!

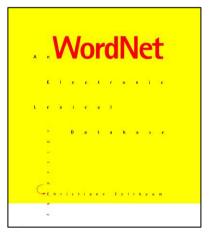
There is only one catch...



A small case study

- Let's try to derive knowledge from texts
- Let's use Boxer to analyse definitions
- Pilot study:

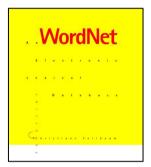
Find axioms for wife and husband





Case study: wife and husband

• WordNet 3.0:



wife (a married woman; a man's partner in marriage)

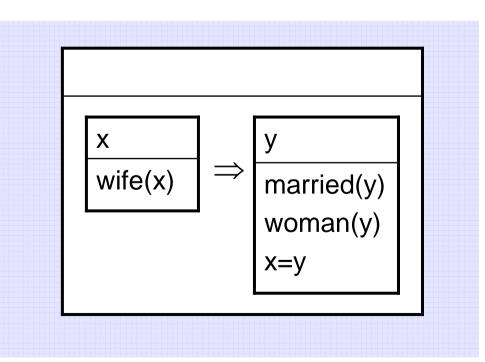
husband (a married man; a woman's partner in marriage)

Boxing WordNet glosses

• Original:

wife (a married woman; a man's partner in marriage)

 Rephrased (1): Every wife is a married woman.

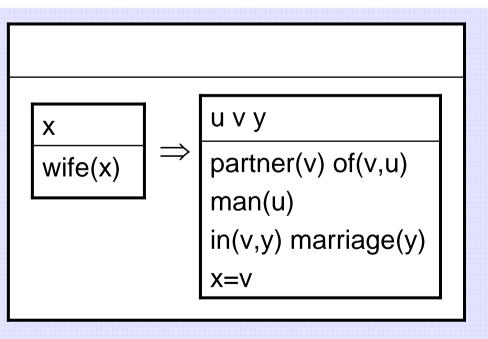


Boxing WordNet glosses

• Original:

wife (a married woman; a man's partner in marriage)

Rephrased (2)
 Every wife is a man's partner in marriage.



Case study: wife and husband

• Wikipedia (Oct 1, 2008):

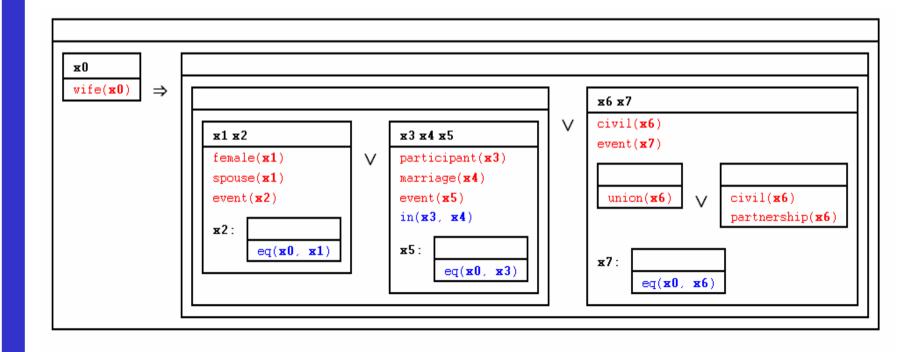


A **wife** is a female spouse, or participant in a marriage, or civil union or civil partnership.

A **husband** is a male spouse (participant) in a marriage, civil union or civil partnership.

Boxing Wikipedia definitions

 Every wife is a female spouse, or participant in a marriage, or civil union or civil partnership.



Boxing glosses and definitions

- Not straightforward to get good results
- (Manual?) reformulation required
- Several issues
 - implication in one or two directions?
 - Word sense disambiguation
 - Modifier attachments
 - Scope of disjunction
 - Interpretation of disjunction
 - Strict or default rules?
- But: inference seems to play no major role

Ideas for definition analysis

- Develop a controlled natural language for definition templates in sources like Wikipedia
- Train new parsing models on annotated definitions

Conclusions

- We can build semantic representations with high coverage and reasonable accuracy for open-domain natural texts
- We also have reasonable inference engines at our disposals that work well enough for small texts
- To use these in real-world applications inference tasks we need additional background knowledge

More Conclusions

- Some of the background knowledge can be derived from existing resources and ontologies
- The word sense problem makes symbol grounding in ontologies hard
- Deriving general knowledge from natural language definitions sounds like a nice idea but is not without stumbles

And finally...

- The C&C tools, Boxer and Nutcracker are freely available for research
 - http://svn.ask.it.usyd.edu.au/trac/candc/wiki
 - There is also on online demo

